

# Tensor Properties of Materials

Bálint Koczor

– 2023 –





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## 1. Basic principles

- Properties and variables
- Scalar and vector variables
- Tensor properties
- Crystal symmetry, Neumann's principle
- Transformation of coordinates, transformation of vectors and tensors
- Representation surface, principal axes

## 2. Second-rank tensors

- Thermal and electrical conductivity
- Electrical and magnetic susceptibility
- Stress and strain
- Thermal expansion
- Optical properties of crystals

## 3. Third and fourth-rank tensors

- Piezoelectricity
- Elastic stiffness and compliance
- Elastic properties of cubic and isotropic crystals

# Linear response in materials

effect		proportionality (property of material)	cause (external)
current density $J$	=	conductivity $\sigma$	electric field $E$
el. polaris. $P$	=	el. suscept. $\epsilon_0 \chi$	electric field $E$
thermal exp. $\epsilon$	=	thermal exp. coeff. $\alpha$	change of temp. $\Delta T$
stress $\sigma$	=	stiffness $C$	strain $\epsilon$

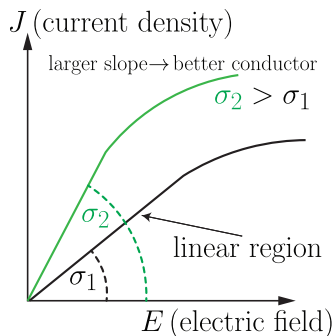
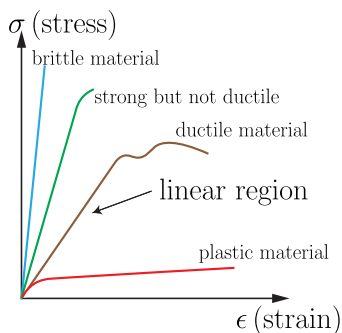
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- valid only in 1D – we need to go beyond numbers  $J$ ,  $P$  etc.
- a physical quantity has both magnitude and direction
- linear relations in vector fields  $\vec{J} = \sigma \vec{E}$

# Linear approximation

deviation from linear: too large current  $\rightarrow$  thermal effects, breakdown

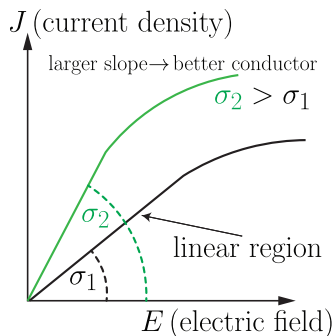
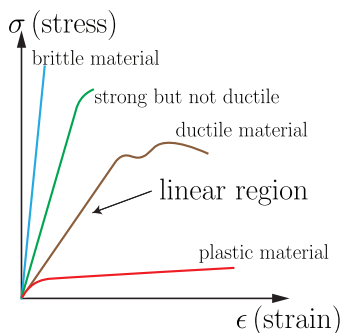


- when cause is 'small' then effect is linear via a Taylor series  

$$f(x) = f(0) + f'(0)x + \dots$$
- slope  $f'(x) = \sigma$  is the proportionality: property of the material

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$$f(x) = f(0) + f'(0)x + \dots$$
- slope  $f'(x) = \sigma$  is the proportionality: property of the material
- in isotropic material  $\sigma$  independent of direction of  $\vec{E}$
- **linear** but anisotropic material: we need to use tensors

## isotropic material

- macroscopic properties  
**independent of direction**
- amorphous materials: glass
- small-grained polycrystalline materials: averaging
- some crystals: linear response in cubic crystals
- proportional vectors  $\vec{J} = \sigma \vec{E}$



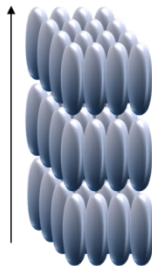


**isotropic material**

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**anisotropic material**

- macroscopic properties **depend on direction**
- liquid crystals
- fibre composites: wood, reinforced concrete
- single crystals: most crystal lattices
- need to write  $\vec{J} = \text{tensor} \cdot \vec{E}$



# Applications of anisotropy

anisotropy is the key to so many technologies and applications

## Piezoelectricity

- quartz oscillators – clock frequency in computers
- gas lighters
- stepper motors and high-precision positioning
- sensors – microphones

## Optical devices

- polarisers
- beam splitters

## Liquid crystals

- LCD screen, displays
- watch
- etc.

# Thought experiment in anisotropic materials

vector  $\vec{J} = (J_x, J_y, J_z)$  not necessarily **parallel** to  $\vec{E} = (E_x, E_y, E_z)$

- we apply electric field  $\vec{E}$  to anisotropic material (**cause**)
- measure the resulting current density vector  $\vec{J}$  (**effect**)
- linearity:  $2\vec{E}$  results in  $2\vec{J}$  with same direction

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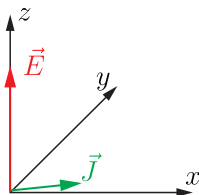
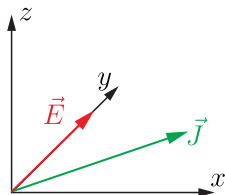
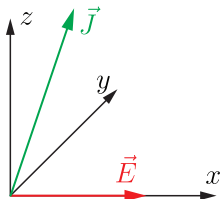
**$x$  direction**

apply  $\vec{E} = E_x \vec{x}$

$$J_x = \sigma_{xx} E_x$$

$$J_y = \sigma_{yx} E_x$$

$$J_z = \sigma_{zx} E_x$$



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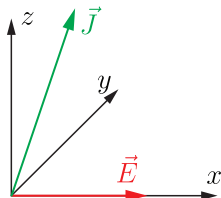
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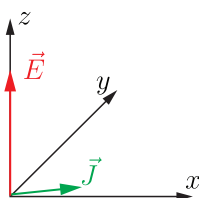
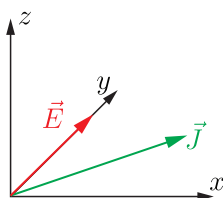
**$y$  direction**

apply  $\vec{E} = E_y \vec{y}$

$$J_x = \sigma_{xy} E_y$$

$$J_y = \sigma_{yy} E_y$$

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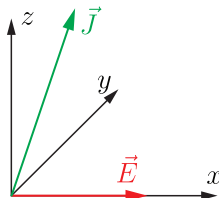
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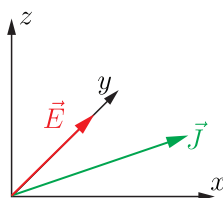
**y direction**

apply  $\vec{E} = E_y \vec{y}$

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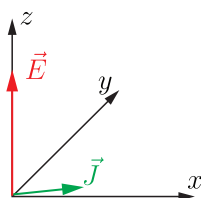
**z direction**

apply  $\vec{E} = E_z \vec{z}$

$$J_x = \sigma_{xz} E_z$$

$$J_y = \sigma_{yz} E_z$$

$$J_z = \sigma_{zz} E_z$$



# Writing linear equations compactly

linearity: if we apply  $\vec{E} = (E_x, E_y, E_z)$  we sum  $x, y, z$  contributions

$$J_x = \sigma_{xx}E_x + \sigma_{xy}E_y + \sigma_{xz}E_z,$$

$$J_y = \sigma_{yx}E_x + \sigma_{yy}E_y + \sigma_{yz}E_z$$

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- notation for axes  $\vec{x} = \vec{x}_1$ ,  $\vec{y} = \vec{x}_2$  and  $\vec{z} = \vec{x}_3$
- and for vector components  $J_x = J_1$ ,  $J_y = J_2$  and  $J_z = J_3$
- and we denote the indexes:  $i, j, k, l \cdots \in \{1, 2, 3\}$



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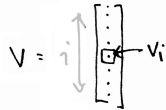
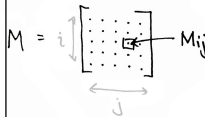
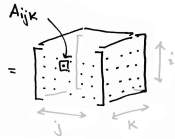
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- and we denote the indexes:  $i, j, k, l \dots \in \{1, 2, 3\}$
- use compact **Einstein convention** or **matrix/vector prod.**

$$J_i = \sum_{j=1}^3 \sigma_{ij} E_j \equiv \sigma_{ij} E_j$$

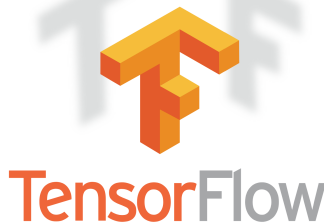
$$\begin{bmatrix} J_1 \\ J_2 \\ J_3 \end{bmatrix} = \begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_{22} & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{bmatrix} \begin{bmatrix} E_1 \\ E_2 \\ E_3 \end{bmatrix}$$

# What is a tensor?

rank- $r$  array of numbers

$r = 0$ scalar: $T$ 	$r = 1$ vector: $V_i$ 	$r = 2$ matrix: $M_{ij}$ 	$r = 3$ (3D) array: $A_{ijk}$ 
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- very useful in so many applications
- materials: express linear properties of materials
- machine learning: tensors are weights in neural networks



# Matrix and vector notation

- for rank  $r = 1, 2$  can use compact **matrix/vector notations**
- for rank  $r \geq 2$  need to use tensors and **Einstein summation**
- Einstein convention: we imply summing over repeated indexes

explicit	Einstein conv.	matrix notation	type
$a = \sum_{i=1}^3 v_i w_i$	$a = v_i w_i$	$a = \vec{v} \cdot \vec{w}$	vector/vector
$a_i = \sum_{j=1}^3 T_{ij} v_j$	$a_i = T_{ij} v_j$	$\vec{a} = \mathbf{T} \cdot \vec{v}$	matrix/vector
$R_{ik} = \sum_{j=1}^3 S_{ij} T_{jk}$	$R_{ik} = S_{ij} T_{jk}$	$\mathbf{R} = \mathbf{S} \cdot \mathbf{T}$	matrix/matrix
$R_{ij} = \sum_{k,l=1}^3 S_{ijkl} T_{kl}$	$R_{ij} = S_{ijkl} T_{kl}$		

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$R_{ij} = \sum_{k,l=1}^3 S_{ijkl} T_{kl}$	$R_{ij} = S_{ijkl} T_{kl}$		

- tensors (as well as matrices/vectors) express linear relations
- apply electric field twice as large  $\vec{E} \rightarrow 2\vec{E}$  then effect  $\vec{J} \rightarrow 2\vec{J}$
- **linearity:** if we make  $\sigma_{ij} \rightarrow 2\sigma_{ij}$  twice as large, same effect

# Summary of lecture 1

## Basic principles

- cause-effect: linear relations used when cause is small
- anisotropy: macroscopic properties depend on direction
- cause vector is not necessarily parallel to effect vector
- in this case we used tensors to express linear relations

## Tensors basics

- tensors, such as  $T_{ijk}$ , are rank- $r$  arrays of numbers
- with materials we work in 3D and indexes run for  $i, j, k = 1, 2, 3$
- rank 1 and rank 2: we can use matrices and vectors
- for higher ranks we use Einstein convention: summation over repeated index is implied



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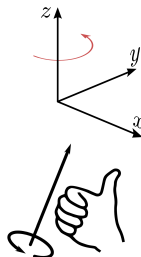
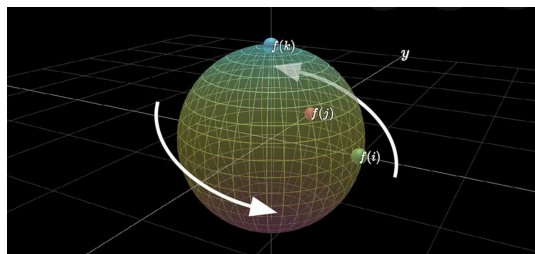
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# Coordinate system rotations

- isotropic materials: rotation does not affect properties
- anisotropic materials: problems simplify when viewed in specific coordinate systems – we want to apply rotations

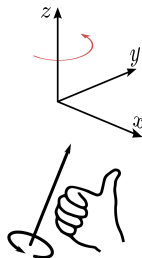
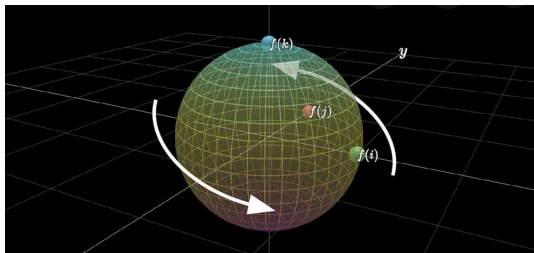


# Coordinate system rotations

- isotropic materials: rotation does not affect properties
- anisotropic materials: problems simplify when viewed in specific coordinate systems – we want to apply rotations
- compactly describe rotations using rotation matrices

$$\vec{J}' = \mathbf{L}\vec{J} \qquad J'_i = L_{ij}J_j$$

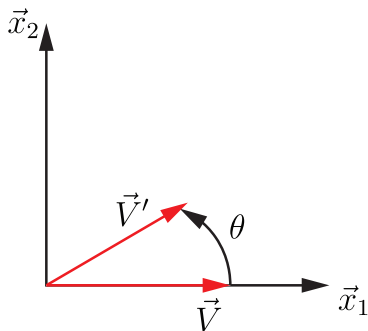
- computer graphics: rotate vectors whenever we move camera





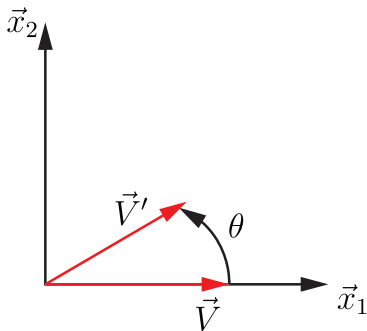
## Example: matrix of rotation around $z$ axis

Rotate vectors that are parallel with coord. axes:  $V_1\vec{x}_1$ ,  $V_2\vec{x}_2$ ,  $V_3\vec{x}_3$



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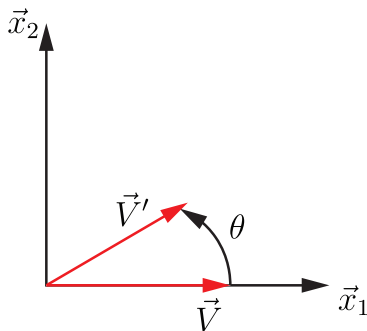
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$$\mathbf{L} \begin{bmatrix} V_1 \\ 0 \\ 0 \end{bmatrix} = V_1 \begin{bmatrix} \cos \theta \\ \sin \theta \\ 0 \end{bmatrix}$$

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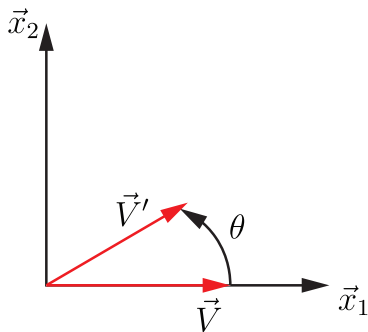


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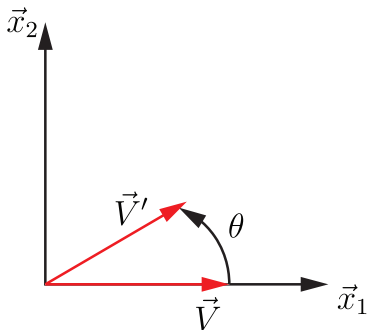
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- a general vector is just the sum  $\vec{V} = V_1\vec{x}_1 + V_2\vec{x}_2 + V_3\vec{x}_3$
- right-hand sides above must be column vectors of  $\mathbf{L}$

$$\begin{bmatrix} V'_1 \\ V'_2 \\ V'_3 \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

# Basic and general rotations

Basic rotation matrices:

$$R_x = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{bmatrix}, R_y = \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}, R_z = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

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But how about an off-axis rotation? General rot. matrix  $\mathbf{L}$

$$\mathbf{L} = \begin{bmatrix} L_{11} & L_{12} & L_{13} \\ L_{21} & L_{22} & L_{23} \\ L_{31} & L_{32} & L_{33} \end{bmatrix}, \quad \mathbf{L} \text{ must be orthogonal and } \det \mathbf{L} = 1$$

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We can use  $\mathbf{L}$  to rotate all axes  $\vec{x}'_i = \mathbf{L}\vec{x}_i$  (see colour in prev. slide)

$$\vec{x}'_1 = \begin{bmatrix} L_{11} \\ L_{21} \\ L_{31} \end{bmatrix}, \quad \vec{x}'_2 = \begin{bmatrix} L_{12} \\ L_{22} \\ L_{32} \end{bmatrix}, \quad \vec{x}'_3 = \begin{bmatrix} L_{13} \\ L_{23} \\ L_{33} \end{bmatrix}$$

- $L_{ij} = \cos \theta_{ij}$  are scalar products between orig. and new axes
- **example:**  $\theta_{12}$  is angle between original  $\vec{x}_1$  and new  $\vec{x}'_2$



# Orthogonal matrices

- But new coordinate axes  $\vec{x}'_i$  must be orthogonal and normalised
- example:  $L_{i1}L_{i1} = \vec{x}'_1 \cdot \vec{x}'_1 = 1$  while  $L_{i1}L_{i2} = \vec{x}'_1 \cdot \vec{x}'_2 = 0$

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- in **L** column and row vectors are mutually orthogonal

$$\boxed{\mathbf{L}^T \mathbf{L} = \mathbb{1} \qquad L_{ij}L_{ik} = \delta_{jk}}$$

- Kronecker delta  $\delta_{ij} = 1$  when  $i = j$  and  $\delta_{ij} = 0$  otherwise
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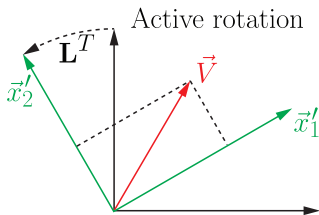
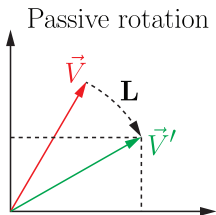
**reminder:** diagonalising matrices via eigenvectors

$$\mathbf{M} = \mathbf{L} \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} \mathbf{L}^T$$

- diagonalise **real, symmetric** matrix **M** with **real eigenvalues  $\lambda_i$**
- here **L** contains eigenvectors of **M** as columns – orthogonal

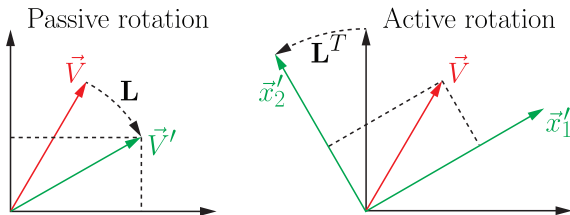
# Active and Passive rotations

- so far, actively rotated vector to a new one  $\vec{J}' = \mathbf{L}\vec{J}$
- physical variable  $\vec{J}$  should not change, just frame of ref
- passive: vector  $\vec{J}$  remains unchanged, entries wrt new coord syst



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**active rotation**  $\vec{V}' = \mathbf{L}\vec{V}$  **equivalent to:** passive rotation with inversely rotated coord syst  $\vec{x}'_i = \mathbf{L}^T \vec{x}_i$

$$\begin{bmatrix} \vec{V}' \cdot \vec{x}_1 \\ \vec{V}' \cdot \vec{x}_2 \\ \vec{V}' \cdot \vec{x}_3 \end{bmatrix} = \begin{bmatrix} V'_1 \\ V'_2 \\ V'_3 \end{bmatrix} = \begin{bmatrix} \vec{V} \cdot \vec{x}'_1 \\ \vec{V} \cdot \vec{x}'_2 \\ \vec{V} \cdot \vec{x}'_3 \end{bmatrix}$$

active rotation  passive rotation

# Rotating tensors

one coordinate system  $\vec{E}$  and  $\vec{J}$ , in other one  $\vec{E}' = \mathbf{L}\vec{E}$  and  $\vec{J}' = \mathbf{L}\vec{J}$

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$$\vec{J} = \boldsymbol{\sigma} \vec{E}$$

substitute  $\vec{J} = \mathbf{L}^T \vec{J}'$  and  $\vec{E} = \mathbf{L}^T \vec{E}'$

$$\mathbf{L}^T \vec{J}' = \boldsymbol{\sigma} \mathbf{L}^T \vec{E}'$$

multiply with  $\mathbf{L}$  from left

$$\mathbf{L} \mathbf{L}^T \vec{J}' = \mathbf{L} \boldsymbol{\sigma} \mathbf{L}^T \vec{E}'$$

simplify  $\mathbf{L} \mathbf{L}^T = \mathbb{1}$

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denote  $\boldsymbol{\sigma}' = \mathbf{L} \boldsymbol{\sigma} \mathbf{L}^T$

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Tensor entries transform due to rotation

$$\boldsymbol{\sigma}' = \mathbf{L} \boldsymbol{\sigma} \mathbf{L}^T \qquad \sigma'_{ij} = L_{ik} \sigma_{kl} L_{jl}$$



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Tensor entries transform due to rotation

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**remark:** formal definition of a rank- $r$  Cartesian tensor

- a rank- $r$  array of real numbers  $T_{i_1 i_2 \dots i_r}$  as before
- transforms according to  $T'_{i_1 i_2 \dots i_r} = L_{i_1 j_1} L_{i_2 j_2} \cdots L_{i_r j_r} T_{j_1 j_2 \dots j_r}$
- when vectors transform according to  $V'_i = \mathbf{L}_{ij} V_j$

# Principal axis system

- given a **symmetric** rank-2 tensor  $\sigma_{ij} = \sigma_{ji}$  as the matrix  $\sigma$
- much simpler to work with tensor in the **principal axis system**
- column vectors of  $\mathbf{L}$  are the eigenvectors  $\vec{x}_i^{pas}$  as **principal axes**

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$$\sigma = \mathbf{L} \begin{bmatrix} \sigma_1^{pas} & 0 & 0 \\ 0 & \sigma_2^{pas} & 0 \\ 0 & 0 & \sigma_3^{pas} \end{bmatrix} \mathbf{L}^T$$

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- parallel **cause** and **effect** in principal directions: eigenvalue eq.

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- parallel **cause** and **effect** in principal directions: eigenvalue eq.

$$\vec{J} = \sigma \vec{E}$$

$$\text{substitute } \vec{E} = \vec{x}_i^{pas}$$

$$\vec{J} = \sigma \vec{x}_i^{pas}$$

$$\text{eigenvalue eq. } \sigma \vec{x}_i^{pas} = \sigma_i^{pas} \vec{x}_i^{pas}$$

$$\vec{J} = \sigma_i^{pas} \vec{x}_i^{pas}$$

$$\text{substitute back } \vec{x}_i^{pas} = \vec{E}$$

$$\vec{J} = \sigma_i^{pas} \vec{E}$$

$$\text{proportionality is a scalar } \sigma_i^{pas}$$

# Summary of lecture 2

## Rotations

- simplify description of anisotropic materials
- rotated axis vectors  $\vec{x}'_i = \mathbf{L}\vec{x}_i$  are column vectors of  $\mathbf{L}$
- active  $\mathbf{L}\vec{v}$  equivalent to inverse passive rot of coord syst
- $\mathbf{L}$  transforms tensors into new coord syst as  $\mathbf{LTL}^T$

## Principal axis system

- can always find a rotation that diagonalises a (symmetric) tensor
- eigenvectors of matrix are the **principal axes**
- eigenvalues of the matrix are the **principal tensor components**
- cause and effect are parallel in the principal directions



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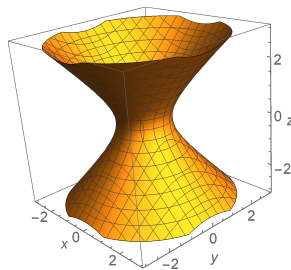
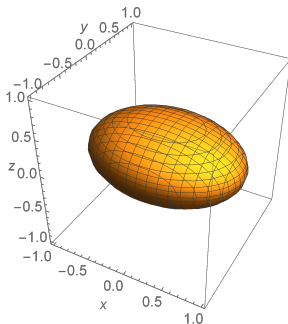
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# Visualisation and crystal symmetries

- rotations are crucial for treating anisotropy
- rotation matrices are very useful for computations
- **but** tensors rather abstract objects  $\rightarrow$  visualisation
- we now introduce intuitive visualisations of tensors

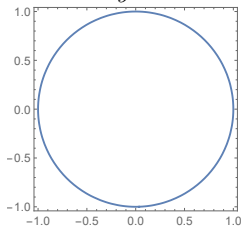


- single crystals have high symmetry
- this manifests in tensor properties

# Reminder: surfaces via implicit equations

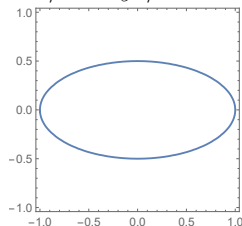
2D circle

$$x^2 + y^2 = 1$$



2D ellipse

$$x^2/a^2 + y^2/b^2 = 1$$

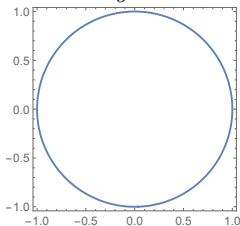




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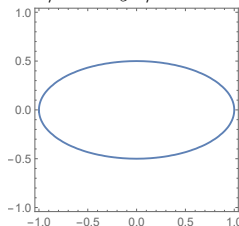
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$$x^2 + y^2 = 1$$



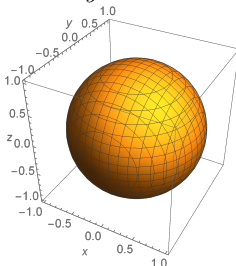
2D ellipse

$$x^2/a^2 + y^2/b^2 = 1$$



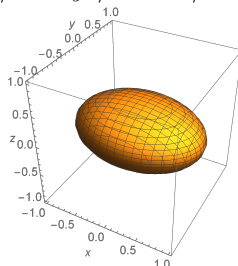
3D sphere

$$x^2 + y^2 + z^2 = 1$$



3D ellipsoid

$$x^2/a^2 + y^2/b^2 + z^2/c^2 = 1$$



# Visualising tensors as surfaces

all  $\vec{x} = (x, y, z)^T$  that produce a parallel component  $\vec{x} \cdot \vec{v} = 1$

$$\text{all } \vec{x}: \quad \vec{x}^T \mathbf{T} \vec{x} = \vec{x} \cdot \underbrace{(\mathbf{T} \vec{x})}_{\vec{v}} = 1 \quad \text{for fixed } \mathbf{T}$$

points  $\vec{x}$  implicitly define a surface in 3D, representative of  $\mathbf{T}$

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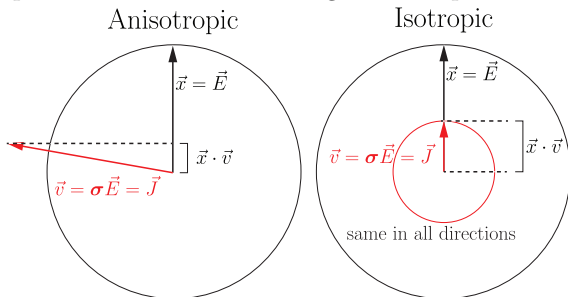
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Example: apply electric field  $\vec{E} = \vec{x}$

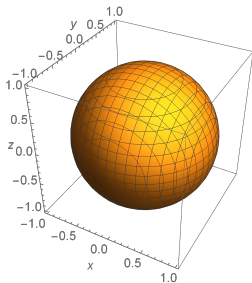
- then vector  $\vec{v} = \sigma \vec{E} = \vec{J}$  is current density
- scalar product  $\vec{x} \cdot \vec{v} = \vec{E} \cdot \vec{J}$  is magnitude of parallel component



# Examples of surfaces

isotropic material:  $T$  is scalar

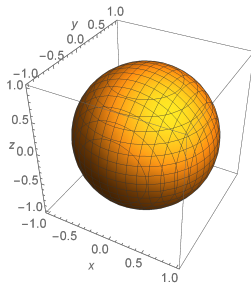
- simplifies  $1 = \vec{x} \cdot (\mathbf{T}\vec{x}) = T\vec{x} \cdot \vec{x}$
- where  $\vec{x} \cdot \vec{x} = x^2 + y^2 + z^2$
- surface:  $1/T = x^2 + y^2 + z^2$
- sphere of radius  $\sqrt{1/T}$



# Examples of surfaces

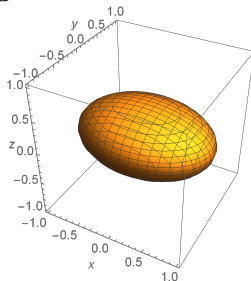
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## Principal Axis System and positive eigenvalues

- $\mathbf{T} = \text{diag}(\lambda_1, \lambda_2, \lambda_3)$
- then  $\vec{v} = \mathbf{T}\vec{x} = (\lambda_1 x, \lambda_2 y, \lambda_3 z)^T$
- ellipsoid:  $1 = \lambda_1 x^2 + \lambda_2 y^2 + \lambda_3 z^2$
- $a = \lambda_1^{-1/2}$ ,  $b = \lambda_2^{-1/2}$ ,  $c = \lambda_3^{-1/2}$



# Representation quadrics

- $T$  is symmetric  $\rightarrow$  eigenvalues  $\lambda_1, \lambda_2, \lambda_3$  are real
- ellipsoid if and only if  $\lambda_1, \lambda_2, \lambda_3 > 0$  are positive
- generalisation of ellipsoid: quadric surfaces in PAS as

$$1 = \lambda_1 x^2 + \lambda_2 y^2 + \lambda_3 z^2$$

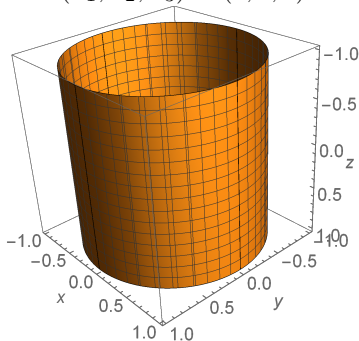
# Representation quadrics

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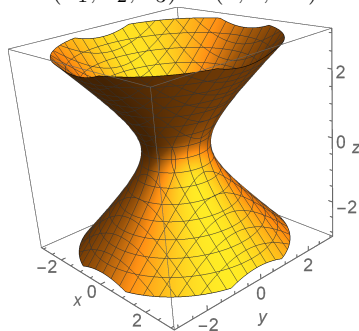
**cylinder**

one component zero  
 $(\lambda_1, \lambda_2, \lambda_3) = (1, 1, 0)$



**hyperboloid**

one component negative  
 $(\lambda_1, \lambda_2, \lambda_3) = (1, 1, -1)$



# Rotational properties

start with: representation quadric of the tensor  $\mathbf{T}$

$$1 = \vec{x}^T \mathbf{T} \vec{x}$$

substitute diagonal  $\mathbf{T} = \mathbf{L} \mathbf{D} \mathbf{L}^T$

$$1 = (\mathbf{L}^T \vec{x})^T \mathbf{D} (\mathbf{L}^T \vec{x})$$

substitute new repr. coords.  $\mathbf{L}^T \vec{x} = \vec{x}'$

$$1 = (\vec{x}')^T \mathbf{D} \vec{x}'$$

this is: representation quadric of diagonal  $\mathbf{D}$

any symm. tensor  $\mathbf{T}$ : take  $\mathbf{T}$  in PAS and rotate with eigenvectors



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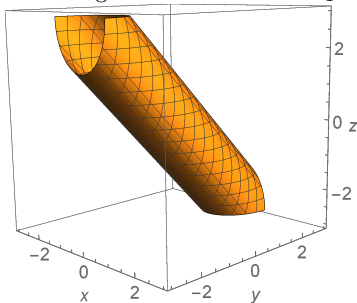
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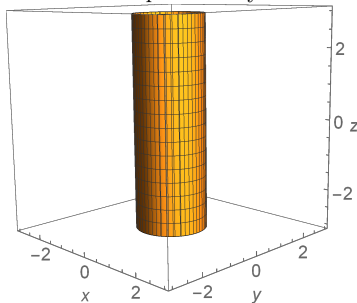
$$\mathbf{T} = \mathbf{L} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{L}^T$$

rotating  $\mathbf{D}$  with  $45^\circ$  around  $\vec{x}_1$



$$\mathbf{D} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

in Principal Axis System



**Neumann's principle:** *Any physical property of a crystal must include the symmetry elements of the point group of the crystal*

- symmetry may be higher, i.e., more invariants, but must include at least those symmetry elements
- symmetry of a rank-2 tensor depends only on crystal system, but not on particular point groups
- representation surface of a tensor also inherits these symmetry components
- crystal symmetry axes determine principal directions, i.e., the principal axis system



triclinic  
(anorthic)

$$a \neq b \neq c \neq a$$

$$\alpha \neq \beta \neq \gamma \neq \alpha$$



monoclinic

$$a \neq b \neq c \neq a$$

$$\alpha = \gamma = 90^\circ$$

$$\beta \neq 90^\circ$$



trigonal  
hexagonal

$$a = b \neq c$$

$$\alpha = \beta = 90^\circ$$

$$\gamma = 120^\circ$$



orthorhombic

$$a \neq b \neq c \neq a$$

$$\alpha = \beta = \gamma = 90^\circ$$



rhombohedral

$$a = b = c$$

$$\alpha = \beta = \gamma \neq 90^\circ$$



tetragonal

$$a = b \neq c$$

$$\alpha = \beta = \gamma = 90^\circ$$



cubic

$$a = b = c$$

$$\alpha = \beta = \gamma = 90^\circ$$

# Effect of crystal structure

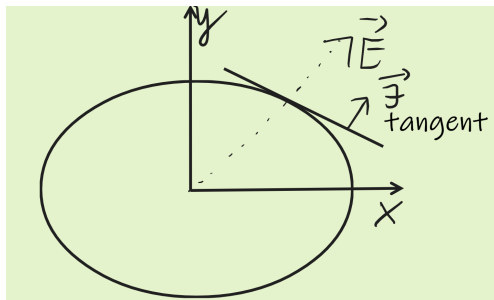
cryst. syst.	quadric orientation	params.	tensor
cubic	sphere	1	$\begin{bmatrix} T & 0 & 0 \\ 0 & T & 0 \\ 0 & 0 & T \end{bmatrix}$
tetragonal, hexagonal, trigonal	symm. around $\vec{x}_3$	2	$\begin{bmatrix} T_1 & 0 & 0 \\ 0 & T_1 & 0 \\ 0 & 0 & T_3 \end{bmatrix}$
orthorhombic	$\vec{x}_1, \vec{x}_2, \vec{x}_3$ parallel to diads	3	$\begin{bmatrix} T_1 & 0 & 0 \\ 0 & T_2 & 0 \\ 0 & 0 & T_3 \end{bmatrix}$
monoclinic	$\vec{x}_2$ parallel to diad	4	$\begin{bmatrix} T_{11} & 0 & T_{13} \\ 0 & T_{22} & 0 \\ T_{13} & 0 & T_{33} \end{bmatrix}$
triclinic	-	6	$\begin{bmatrix} T_{11} & T_{12} & T_{13} \\ T_{12} & T_{22} & T_{23} \\ T_{13} & T_{23} & T_{33} \end{bmatrix}$

- monoclinic: 3 eigenvalues + 1 angle
- triclinic: 3 eigenvalues + 3 angles
- others: crystal axes define tensor PAS

# Radius normal property

- recall the representation quadric surface  $1 = \vec{x}^T \boldsymbol{\sigma} \vec{x}$
- in PAS it is  $1 = \sigma_1 x^2 + \sigma_2 y^2 + \sigma_3 z^2$  with conductivities  $\sigma_1, \sigma_2, \sigma_3$
- the tangent plane vector is  $\vec{t}(\vec{x}) = (2\sigma_1 x, 2\sigma_2 y, 2\sigma_3 z)^T$
- we compute as  $\vec{t} = \vec{\nabla} F(x, y, z)$  where  $0 = F(x, y, z) = \vec{x}^T \boldsymbol{\sigma} \vec{x} - 1$
- statement:  $\vec{J} = \boldsymbol{\sigma} \vec{E}$  is parallel with  $\vec{t}(\vec{E})$

$$\vec{J} = (\sigma_1 E_1, \sigma_2 E_2, \sigma_3 E_3)^T \quad \vec{t}(\vec{E}) = 2(\sigma_1 E_1, \sigma_2 E_2, \sigma_3 E_3)^T$$



# Summary of lecture 3

## Visualisation of tensors

- applicable to symmetric rank-2 tensors
- surface: collection of points that satisfy quadric equation
- depending on eigenvalues of tensor we get different quadrics
- quadric surfaces: sphere, ellipsoid, hyperboloid, cylinder etc.
- quadrics transform naturally when transforming tensors

## Effect of crystal structure

- Neumann's principle: physical properties of crystals must include cryst. symmetries
- therefore tensors include crystal symmetries via eigenvalues and eigenvectors
- these symmetries are also included by representation quadrics



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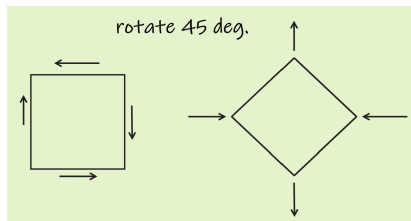
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# Applications of tensors

- now equipped with mathematical tools
- illustrated concepts on linear response in anisotropic materials
- there tensors expressed properties of materials
- example  $\vec{J} = \sigma \vec{E}$

## some more advanced applications of tensors

- mechanical stress and strain: tensor no longer property
- thermal expansion: cause is scalar
- stiffness via Hooke's law: tensor no longer rank 2

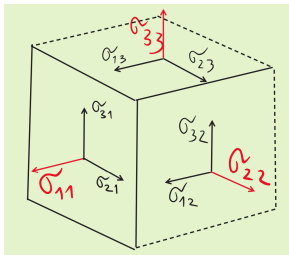


# Stress tensor

- take any infinitesimally small cube of material
- traction vector  $\vec{T}$  is force per unit area
- that acts on the plane represented by area vector  $\vec{n}$

$$\vec{T} = \boldsymbol{\sigma} \vec{n} \qquad T_i = \sigma_{ij} n_j$$

- above guarantees rotational property  $\boldsymbol{\sigma}' = \mathbf{L} \boldsymbol{\sigma} \mathbf{L}^T$  via Slide 19
- stress tensor is not a property of material: rather a variable
- equilibrium: tensile stresses equal but opposite  $\boldsymbol{\sigma}(-\vec{n}) = -\vec{T}$



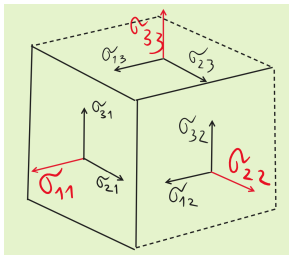


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**Warning:** row vector convention

- often **row vectors** as  $\vec{T}^T = \vec{n}^T \boldsymbol{\sigma}_{row}$
- need to use transpose  $\boldsymbol{\sigma}_{row} = \boldsymbol{\sigma}^T$
- need to use inverse rotation  $\mathbf{L}^T$  for row vectors
- possible confusion:  $\mathbf{L}^T$  is inverse rotation in our convention

# Equilibrium: conservation of angular momentum

- decompose any matrix into symm. and antisymm.  $\sigma = \sigma^s + \sigma^a$
- we can define  $2\sigma^s = \sigma + \sigma^T$  and  $2\sigma^a = \sigma - \sigma^T$
- **Cauchy:** conservation of ang. mom.  $\sigma \equiv \sigma^s$  is symmetric

$$\begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_{22} & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{bmatrix} =$$

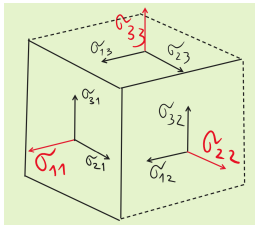
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positive: tensile

negative: compressive



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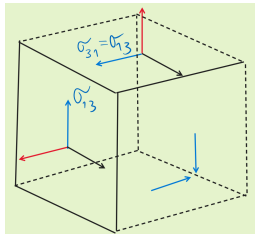
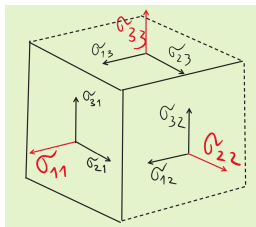
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shear stress

negative: compressive conserves ang. mom.



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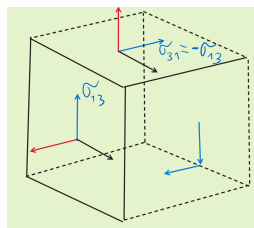
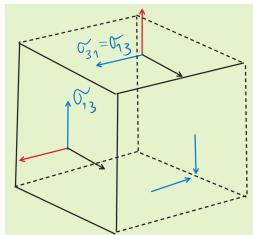
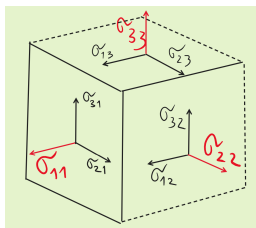
negative: compressive

shear stress

conserves ang. mom.

rotation of the cube

zero in equilibrium



# Principal stresses

- in equilibrium we can diagonalise  $\sigma = \sigma^T$
- but  $\sigma$  is not property of material: eigenvectors not necessarily related to crystal axes

stress type	principal comps.	example	tensor
triaxial	3 non-zero $\sigma_k$	-	$\text{diag}(\sigma_1, \sigma_2, \sigma_3)$
biaxial	2 non-zero $\sigma_k$	force on thin plate	$\text{diag}(\sigma_1, \sigma_2, 0)$
uniaxial	1 non-zero $\sigma_k$	pulling wire	$\text{diag}(\sigma_1, 0, 0)$
hydrostatic	3 identical $\sigma_k < 0$	pressure $p$ in fluid	$\text{diag}(-p, -p, -p)$
pure shear	special biaxial	rod torsion	$\text{diag}(-\sigma, \sigma, 0)$

# Principal stresses

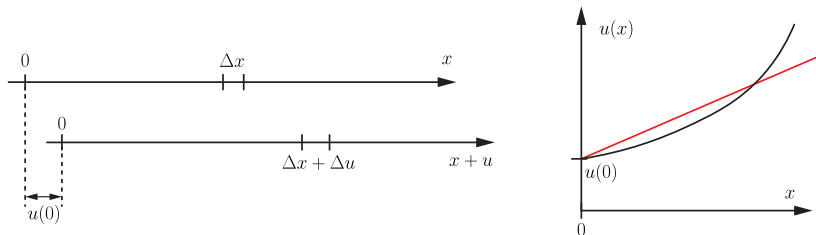
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**remark:** how 'big' is the stress? no obvious metric as with vectors  
matrix invariants:  $\text{Tr}(A)$ , determinant and matrix norms as  $\text{Tr}(A^T A)$

# Strain in 1D

length of small element  $\Delta x$  increases due to stretching as  $\Delta x + \Delta u$



$$\text{strain} = \frac{\text{increase in length}}{\text{original length}} = \lim_{\Delta x \rightarrow 0} \frac{\Delta u}{\Delta x} = \frac{du}{dx}$$

- small element  $(x, x + \Delta x)$  is increased to  $(x + u, x + \Delta x + u + \Delta u)$
- strain  $\epsilon$  is the derivative of displacement  $u(x)$  wrt. position  $x$
- **homogeneous**: derivative constant, global property
- Taylor expansion in homogeneous case:  $u(x) - u(0) = \epsilon x$



# Strain in 3D

- need to use vectors  $\vec{u} = (u_1, u_2, u_3)^T$  and  $\vec{x} = (x_1, x_2, x_3)^T$
- tensor: partial derivatives of displacement field wrt position

$$\tilde{\epsilon}_{ij} = \frac{\partial u_i}{\partial x_j}$$

- as before, antisymmetric part  $(\tilde{\epsilon}_{ij} - \tilde{\epsilon}_{ji})/2$  is rotation

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- strain is a symmetric, rank-2 tensor

$$\epsilon_{ij} = \left( \frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right) / 2$$

$$\epsilon = \underbrace{\begin{bmatrix} \epsilon_{11} & 0 & 0 \\ 0 & \epsilon_{22} & 0 \\ 0 & 0 & \epsilon_{33} \end{bmatrix}}_{\text{uniaxial extension}} + \underbrace{\begin{bmatrix} 0 & \epsilon_{12} & \epsilon_{13} \\ \epsilon_{12} & 0 & \epsilon_{23} \\ \epsilon_{13} & \epsilon_{23} & 0 \end{bmatrix}}_{\text{shear}}$$

- diagonals  $\epsilon_{11}, \epsilon_{22}, \epsilon_{33}$  are uniaxial extensions per unit length
- off-diagonals are shear strains – coordinate syst. dependent
- homogeneous case:  $\vec{u}(\vec{x}) - \vec{u}(0) = \epsilon \vec{x}$

# Pure shear stress and strain

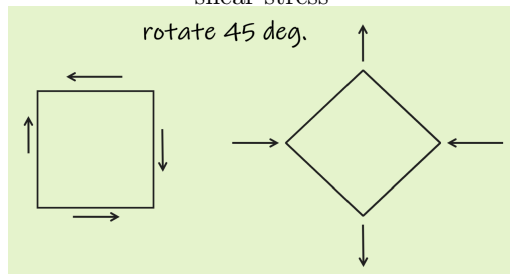
assume the pure shear stress and strain tensors

$$\sigma = \begin{bmatrix} 0 & \sigma & 0 \\ \sigma & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \epsilon = \begin{bmatrix} 0 & \epsilon & 0 \\ \epsilon & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

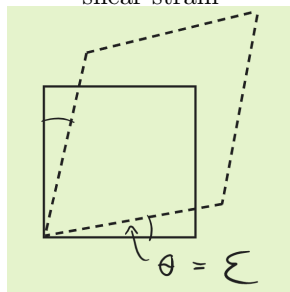
- both can be diagonalised by a  $45^\circ$  rotation of the coord. syst.
- two eigenvalues  $\pm\sigma$  and  $\pm\epsilon$
- for strain  $\epsilon = \epsilon_{12} = (\frac{\partial u_1}{\partial x_2} + \frac{\partial u_2}{\partial x_1})/2 = \tan(\theta) \approx \theta$

shear stress

rotate 45 deg.



shear strain



# Strain tensor via thermal expansion

- material is heated up uniformly by  $\Delta T$
- strain tensor is proportional to this temperature
- proportionality: symmetric thermal expansion coefficients  $\alpha_{ij}$

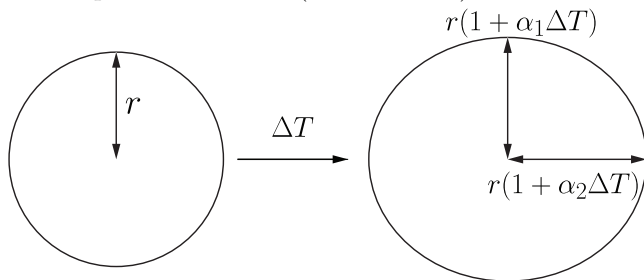
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- eigenvalues and eigenvectors contain crystal symmetries
- principal components  $\epsilon_i = \alpha_i \Delta T$ , e.g., along crystal axes
- $\alpha_i$  are typically positive, but can be negative
- volume expansion is  $\Delta V \approx (\alpha_1 + \alpha_2 + \alpha_3)V\Delta T$  an invariant



# Hooke's law and linear elasticity

in case of a 1D spring we have Hooke's law

$$F = kx$$

- can relate stress and strain in an elastic material
- linear mapping via rank-4 stiffness tensor  $c_{ijkl}$

$$\sigma_{ij} = c_{ijkl} \epsilon_{kl}$$

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- $c_{ijkl}$  may generally have  $3^4 = 81$  entries, but reduces to 21
- example:  $\sigma_{ij} = \sigma_{ji}$ , therefore  $c_{ijkl} = c_{jikl}$
- example:  $\epsilon_{kl} = \epsilon_{lk}$ , therefore  $c_{ijkl} = c_{ijlk}$
- in crystals this further reduces due to symmetries in  $\epsilon_{kl}$
- orthorhombic crystals: 9 entries, hexagonal crystals: 5, cubic: 3